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COURSE BROCHURE

Digital Twin and AI Applications in Pipeline Integrity Training Course

26 – 30 April 2027
Casablanca

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Registered in the United Kingdom and Saudi Arabia

Digital Twin and AI Applications in Pipeline Integrity Training Course

Ref: 1313_49257 **Date:** 26 – 30 April 2027 **Location:** Casablanca **Fees:** 4900 GBP

Digital Twin and AI Applications in Pipeline Integrity is a 1-week training course for pipeline integrity engineers and asset managers who decide where to dig, repair, inspect or lower pressure, and who leave with a pipeline integrity twin blueprint. Ageing hydrocarbon and water networks now produce inline inspection, cathodic protection and SCADA data faster than people can read it. Jupiter Prime Training delivers this course on Digital Twin and AI Applications in Pipeline Integrity.

Why Digital Twin and AI Applications in Pipeline Integrity Matter Now

Three pressures meet on every pipeline system: assets running beyond their original design life, regulators and communities expecting fewer releases and faster leak response, and inspection budgets that cannot cover every kilometre each cycle. Operators that fuse inspection, sensor and environmental data into a living model target digs precisely and catch growing defects early; those that rely on static spreadsheets risk ruptures, product loss and costly emergency repairs. The course frames model oversight with ISO/IEC 42001, the international standard for an AI management system, and the NIST AI Risk Management Framework NIST AI RMF, so machine learning predictions remain traceable to engineering judgement.

Who Gains Most from Digital Twin and AI Applications in Pipeline Integrity

- Integrity engineers who assess inline inspection anomalies and set repair priorities
- Corrosion specialists who manage cathodic protection surveys and coating condition
- Asset managers who balance inspection budgets against risk across a network
- Control room and leak detection leads who tune alarms and respond to pressure events
- Data and GIS specialists who maintain pipeline alignment, attributes and inspection histories

A joint delegation of integrity, corrosion and control room staff from one operator gains most, because a pipeline twin only works when those three data owners agree on one shared model.

Digital Twin and AI Applications in Pipeline Integrity Capabilities

Participants move from scattered inspection files and sensor feeds to a connected pipeline model that ranks threats and guides field decisions.

- Design the data architecture for a pipeline digital twin — integrity and IT managers
- Align inline inspection runs and close-interval surveys to a common reference — integrity engineers
- Build a machine learning model for external corrosion growth rates — repair prioritisation
- Test leak detection models against false alarm rates — control room supervisors
- Calculate risk scores that combine likelihood and consequence — inspection budget holders

- Specify a risk-based inspection schedule supported by twin outputs — asset owners and auditors

Key Digital Twin and AI Applications in Pipeline Integrity Terms

- Pipeline Digital Twin — a continuously updated virtual model linking a pipeline's geometry, condition, operating data and threats to support integrity decisions.
- Inline Inspection ILI — internal tool runs using magnetic flux or ultrasonic sensors to detect metal loss, dents and cracks.
- Run-to-Run Alignment — matching anomalies from successive inspection runs so their growth over time can be measured reliably.
- Cathodic Protection — an electrochemical method that applies current to a buried pipeline so its steel surface resists external corrosion.
- Computational Pipeline Monitoring — software that analyses flow, pressure and temperature to infer and locate leaks in real time.
- Risk-Based Inspection — planning inspection type and frequency according to the likelihood and consequence of failure for each segment.

Digital Twin and AI Applications in Pipeline Integrity Day by Day

Day 1: Pipeline Threats and Data Landscape

- Threat Categories for Buried Lines — Corrosion, Cracking and Third-Party Damage
- Integrity Data Inventory — ILI, CP Surveys, SCADA and GIS Layers
- Digital Twin Maturity Levels — From Static Records to Live Models
- Oil, Gas and Water Network Cases — Different Consequence Profiles
- Twin Value Hypothesis — Linking Data Investment to Integrity Decisions

Day 2: Building and Fusing the Twin

- Pipeline Data Model and Linear Referencing — Common Location Framework
- ILI Run-to-Run Alignment — Measuring Defect Growth Reliably
- Cathodic Protection and Coating Data Fusion — External Corrosion Context
- Sensor and SCADA Streams — Pressure Cycling and Fatigue Inputs
- Data Quality Rules — Gaps, Offsets and Conflicting Records

Day 3: Machine Learning for Corrosion and Leaks

- Corrosion Growth Regression Models — Predicting Time to Critical Depth
- Soil, Terrain and Environmental Features — Explaining Corrosion Hotspots
- Leak Detection with Anomaly Models — Pressure Transients and Mass Balance
- Acoustic and Fibre Optic Sensing Analytics — Locating Small Releases
- Model Validation Against Dig Results — Checks Aligned with NIST AI RMF

Day 4: Risk-Based Inspection and Decisions

- Likelihood and Consequence Scoring — Segment Risk Ranking
- Fitness-for-Service Screening — Dig, Repair or Monitor Choices
- Inspection Interval Optimisation — Budget Versus Risk Trade-Offs
- Pressure Derating Scenarios — Throughput Impact of Integrity Limits
- Integrity Dashboards for Management — Communicating Risk Movement

Day 5: Practising the Pipeline Integrity Twin Blueprint

- Data Architecture Section — Applying Day 2 Fusion and Referencing
- Prediction Use Cases — Reusing Day 3 Corrosion and Leak Models
- Inspection Strategy Section — Drawing on Day 4 Risk Ranking
- Model Accountability Controls Under ISO/IEC 42001 — Ownership and Revalidation
- Pipeline Integrity Twin Blueprint — Phasing, Costs and Management Sign-Off

Applying Digital Twin and AI Applications in Pipeline Integrity After the Course

- Align the two most recent inline inspection runs on one segment and report the ten fastest-growing anomalies.
- Compare current leak detection alarm logs with confirmed events and propose threshold changes to the control room.

Digital Twin and AI Applications in Pipeline Integrity Questions FAQ

What are Digital Twin and AI Applications in Pipeline Integrity and why do they matter now?

They combine a live virtual pipeline model with machine learning to predict corrosion, cracking and leaks. They matter now because networks are ageing, inspection budgets are stretched and stakeholders expect fewer releases and faster response.

Which roles gain most from Digital Twin and AI Applications in Pipeline Integrity?

Integrity engineers, corrosion specialists, asset managers, leak detection leads and GIS or data specialists gain most. Each owns part of the inspection, sensor or location data that a pipeline twin must bring together.

How do Digital Twin and AI Applications in Pipeline Integrity differ from a general equipment digital twin course?

The focus is on long buried linear assets rather than plant machinery. General equipment twin courses model rotating or static plant items; this one centres on inline inspection alignment, cathodic protection, leak detection and segment risk ranking.

What do you bring back from Digital Twin and AI Applications in Pipeline Integrity?

You bring back a phased design for your network's integrity model. It covers data architecture, prediction use cases, inspection strategy and oversight controls, ready for management review and budgeting.

Digital Twin and AI Applications in Pipeline Integrity: The Way Forward

The pipeline integrity twin blueprint supports the first decision an operator faces: which segment to model first. Integrity and control room teams use it to agree data owners and priorities. After the first segment goes live, revisit the corrosion growth assumptions, because early dig results usually show where models over- or under-predict.